



Evaluating genetic traceability methods for captive-bred marine fish and their applications in fisheries management and wildlife forensics

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ABSTRACT: Growing demands for marine fish products is leading to increased pressure on already depleted wild populations and a rise in aquaculture production. Consequently, more captive-bred fish are released into the wild through accidental escape or deliberate releases. The increased mixing of captive-bred and wild fish may affect the ecological and/or genetic integrity of wild fish populations. Unambiguous identification tools for captive-bred fish will be highly valuable to manage risks (fisheries management) and tracing of escapees and seafood products (wildlife forensics). Using single nucleotide polymorphism (SNP) data from captive-bred and wild populations of Atlantic cod *Gadus morhua* L. and sole *Solea solea* L., we explored the efficiency of population and parentage assignment techniques for the identification and tracing of captive-bred fish. Simulated and empirical data were used to correct for stochastic genetic effects. Overall, parentage assignment performed well when a large effective population size characterized the broodstock and escapees originated from early generations of captive breeding. Consequently, parentage assignments are particularly useful from a fisheries management perspective to monitor the effects of deliberate releases of captive-bred fish on wild populations. Population assignment proved to be more efficient after several generations of captive breeding, which makes it a useful method in forensic applications for well-established aquaculture species. We suggest the implementation of a case-by-case strategy when choosing the best method.

KEY WORDS: Aquaculture · Conservation genetics · Escapees · Fisheries management · Wildlife forensics

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INTRODUCTION

Aquaculture is one of the fastest growing food-producing sectors and will remain so in the foreseeable future due to a growing human demand for animal protein and lipids (Braithwaite & Salvanes 2010) and the limits that have been reached for wild-capture fisheries production (FAO Fisheries and Aquaculture Department 2014). This has led to various challenges related to the aquaculture industry, including organic, chemical and pharmaceutical pollution (Seymour & Bergheim 1991), infectious diseases (Murray & Peeler 2005), feed supply (Naylor et al. 2000, 2009, Natale et al. 2013) and escapees (Kitada et al. 2009, Glover 2010, Glover et al. 2011, Noble et al. 2014).

Accidental escapees (Bekkevold et al. 2006, Glover et al. 2013, Noble et al. 2014) or deliberate releases (Bell et al. 2008, Kitada et al. 2009) of captive-bred marine fish may impact the environment, and the ecological and genetic integrity of wild fish populations (Braithwaite & Salvanes 2010, Laikre et al. 2010). First, a decrease in genetic diversity, and consequently a lower evolutionary potential, has been observed in wild marine fish populations which have been invaded by captive-bred conspecifics (Hindar et al. 1991, Weir & Grant 2005, Glover et al. 2013). Given that recent studies have indicated surprisingly fine-scale local genetic adaptation in marine fish (André et al. 2011, Nielsen et al. 2012, Vandamme et al. 2014), the introgression of captive-bred fish can be detrimental to the long-term survival of wild fish populations. Second, introgression might disrupt adaptive gene complexes, which reduces the fitness of hybrids and in turn may compromise the persistence of locally adapted populations (McGinnity et al. 2003, Danancher & Garcia-Vazquez 2011, Lamaze et al. 2013). Managing and mitigating risks and assessing the impacts of released/escaped captive-bred fish on local wild populations are thus of utmost importance to ensure the long-term sustainability of aquaculture and fisheries industries. Third, aquaculture companies might have legal obligations to report escapees and failure to comply with these regulations might result in fines (Glover 2010). As such, the ability to trace back escapees to the farm of origin constitutes a highly valuable asset in delivering evidence for legal action (Glover et al. 2008, Glover 2010). Finally, an increase in international trade and consumer awareness in recent decades has highlighted the need for accurate labelling of seafood products. Mislabelling to increase profits has been extensively documented in the seafood industry

(Jacquet & Pauly 2008, Hanner et al. 2011, Mariani et al. 2014). Given that market prices of wild-caught marine fish species are generally higher than aquaculture sourced fish, fraudulent labelling captive-bred fish as 'wild-caught' may increase income for the perpetrator (Cline 2012, Warner et al. 2013). Hence, genetic identification methods for farmed and wild marine fish species would be extremely valuable in aquaculture and fisheries management and wildlife forensics.

For a large variety of commercially reared species, escapees and deliberate releases have been reported (Liao et al. 2003, Bell et al. 2008, Jensen et al. 2010, Danancher & Garcia-Vazquez 2011). However, due to their long breeding history and the availability of genetic tools, research on tracing and quantifying escapees has focused mainly on salmonids (Glover 2010, Glover et al. 2013) and only recently on sea bass and sea bream (Arechavala-Lopez et al. 2013, Somarakis et al. 2013, Brown et al. 2015). Extending standardized traceability methods to other commercially exploited marine fish species will thus advance research into the effects of escapees and restocking programmes.

The lack of a long breeding history in most cultured marine fish species complicates the genetic discrimination between wild and captive-bred marine fish, especially when the identification of the hatchery of origin is required. The recent domestication history of many marine fish results in similar allele frequencies in captive-bred and wild populations, which lowers the discrimination power of genetic markers (Duarte et al. 2007). Likewise, the absence of long-term selective breeding programmes reduces the likelihood of finding species-specific 'domestication' markers (Karlsson et al. 2011, Gjerdem et al. 2012). Stochastic and selective breeding processes in aquaculture and recent developments in genetic traceability tools can however facilitate discrimination between captive-bred and wild fish. The common use of a relatively small broodstock and the unwanted high variance in reproductive success within the hatchery will result in increased genetic differentiation between captive-bred and wild populations and a lower genetic diversity within the captive-bred population (Porta et al. 2006a,b, 2007). Within the marine environment, provided that a solid genetic baseline is available, wild fish can be individually assigned to their region and/or population of origin with high precision using gene-associated single nucleotide polymorphism (SNP) markers (Nielsen et al. 2012). Genetic background information is increasingly available for commercially important

fish species (Nielsen et al. 2009, Abadía-Cardoso et al. 2013, Clemento et al. 2014), which makes the use of simulation studies possible to assess the discrimination power of existing genetic markers for wild and captive-bred fish. Finally, while the rate of genetic drift at neutral markers depends on the effective population size (N_e) of the broodstock, SNP markers associated with important aquaculture traits (such as growth and disease resistance) are subjected to directional selection which will increase the degree of differentiation between wild and captive-bred populations (Glover et al. 2010). Such markers may introgress at different rates compared to selectively neutral markers (Lamaze et al. 2012, Hohenlohe et al. 2013), thus providing crucial insights into both the fitness and molecular consequences of escapees and restocking programmes.

Multiple approaches are available for identifying and discriminating between captive-bred and wild marine fish (Manel et al. 2005). The 2 main methods used to date are individual assignment (IA) and parentage-based tagging (PBT) (Manel et al. 2005, Jones et al. 2010). Most commonly used, IA methods rely on allele frequency differences between populations to assign an individual to its most likely source (Ogden 2008, Glover 2010, Nielsen et al. 2012). However, in order to achieve highly robust assignments, IA requires some level of genetic differentiation between populations and extensive genetic reference data (Manel et al. 2005, Nielsen et al. 2012). In contrast, PBT utilizes the genetic variation within the complete data to determine the most likely parental pair for a particular genotype and can achieve high assignment success even when genetic differentiation among populations is insufficient for IA (Jones & Ardren 2003, Steele et al. 2013).

Our study focuses on Atlantic cod *Gadus morhua* L., 1758 and sole *Solea solea* L., 1958, 2 commercially important fish of the Northeast Atlantic Ocean for which extensive genetic resources are available (Nielsen et al. 2012). Both species have a widespread distribution across the Northeast Atlantic Ocean, and their high commercial value has resulted in an increased interest in captive-breeding programmes, restocking, stock enhancement and sea ranching (Howell 1997, Kjesbu et al. 2006, Björnsson 2011). More specifically, declines in wild-caught Atlantic cod and advances in captive breeding and feed formulation have led to an increase in global aquaculture production, reaching 22 000 tons in 2010 (Rosenlund & Skretting 2006, Thurstan et al. 2010, FAO Fisheries and Aquaculture Department 2015a). Although cod aquaculture has recently decreased

due to large catches on the northern fishing grounds (FAO Fisheries and Aquaculture Department 2015a), the use of traditional cage farming in cod aquaculture and the substantial interest in stock enhancement and sea ranching programmes continues to represent a significant risk for interactions between wild and hatchery-reared cod (Bekkevold et al. 2006, Jørstad et al. 2008, Björnsson 2011). Similarly, recent advances and changing economic perspective have increased the interest in sole aquaculture, with production peaking at 125 tons in 2010 but decreasing in recent years (Howell 1997, Imsland et al. 2003, FAO Fisheries and Aquaculture Department 2015b). Although intensive land-based recirculation systems are currently preferred in flatfish aquaculture, there is considerable interest to reduce production costs through less intensive systems (e.g. cage farming, stock enhancement and sea ranching) (Brown 2002, Kitada & Kishino 2006, Sparrevohn & Støttrup 2007). Hence, for both focal species, there is a considerable risk of introgression between captive-bred individuals and local wild populations.

In this study, we aimed to evaluate the utility of IA and PBT approaches to discriminate between captive-bred marine fish and natural fish populations. To achieve this, we used a combination of simulated and empirical SNP datasets to perform a series of assignment experiments in each species, across a range of potential scenarios. The level of genetic differentiation between captive-bred and wild marine fish will vary due to: (1) the number of captive-bred generations (F_n) prior to escape or release, (2) the number of broodstock and the strength of reproductive variance between broodstock individuals, which both influence N_e , and (3) genetic (and geographical) differences between the hatchery population and locally occurring wild populations with which the escapees will intermingle. We investigated each of these potential variables to evaluate their relative impact on assignment power under IA and PBT approaches. In addition, the effect of (in)complete reference samples was also assessed given that the availability and representative nature of reference samples will also affect traceability outcomes.

From the outset, we anticipated that increasing F_n , decreasing N_e and a distinct genetic origin of the broodstock will all favour IA, given that IA relies on the realized level of genetic differentiation between populations to make robust assignments. On the other hand, the performance of PBT will be negatively impacted by those parameters that reduce the genetic diversity within the captive-bred population (i.e. high F_n and low N_e) due to the difficulty of

excluding candidate parents from real parents. Therefore, in addition to evaluating the relative performance of the 2 approaches, we were interested in examining possible thresholds of F_n and N_e across which the optimal approach for determining fish origin actually changes.

MATERIALS AND METHODS

Sampling

Wild samples of 10 Atlantic cod and 14 sole populations have been previously collected from European waters and genotyped (Nielsen et al. 2012). An Atlantic cod broodstock ($A_{\text{cod-BS}}$) ($n = 92$) sourced from the ICES region 27.V.b2 – Faroe Bank was sampled from the Fiskaaling aquaculture research station (Faroe Islands). Atlantic-sourced (ICES 27.IV.c – Southern North Sea) sole were sampled from a Dutch experimental breeding farm, SOLEA in IJmuiden, and consisted of 2 full-sib families with 4 broodstock individuals ($A_{\text{sole-BS}}$) ($n = 4$) and their first-generation offspring ($F_n = 1$) ($A_{\text{sole-F1}}$) ($n = 92$) (Blonk et al. 2009). Captive-bred sole samples originating from the Mediterranean Sea (FAO 37.2.1 – North Adriatic) were obtained from a pilot farm of the Laboratory of Aquaculture, Department of Veterinary Medical Sciences of the University of Bologna, Italy, and included samples from a broodstock ($M_{\text{sole-BS}}$) ($n = 26$) and first-generation offspring ($F_n = 1$) ($M_{\text{sole-F1}}$) ($n = 96$), obtained from 4 batch spawnings ($M_{\text{sole-F1-B1}}$, $M_{\text{sole-F1-B2}}$, $M_{\text{sole-F1-B3}}$ and $M_{\text{sole-F1-B4}}$). More details on all populations used in this study are found in Supplement 1 (www.int-res.com/articles/suppl/q008p131_supp.pdf).

Genotypic data

Gene-associated SNP markers were available for: the wild populations of Atlantic cod (1258 SNPs), the wild populations of sole (427 SNPs), $A_{\text{cod-BS}}$ (427 SNPs), $A_{\text{sole-BS}}$ and $A_{\text{sole-F1}}$ (423 SNPs) (Table 1) (Nielsen et al. 2012, Diopere et al. 2014). Additional genotyping of the $M_{\text{sole-BS}}$ and $M_{\text{sole-F1}}$ samples was conducted using VeraCode™ technology on the BeadExpress platform (Illumina), following the manufacturer's instructions. Of the 427 available SNPs, the 192 most informative SNPs, showing high genetic discrimination values (F_{ST} values) between the Mediterranean populations, were genotyped (Nielsen et al. 2012). Quality assessment and genotype calling

was performed using GenomeStudio v.2009.2 software (Illumina). Three individuals from the $A_{\text{sole-BS}}$, initially genotyped with the wild populations using the SAM assay (GoldenGate, GG) on the iScan platform (Illumina) and with the highest GG call rate for the selected panel (Diopere et al. 2014), were included as cross-platform genotyping controls to ensure comparability between the archived and newly generated data.

Marker selection

In order to obtain marker panels with sufficient assignment power and to ensure that they are easily transferrable between laboratories, a subset of 96 highly informative SNPs were selected based on the practical limitations of common genotyping platforms (Supplement 2 at www.int-res.com/articles/suppl/q008p131_supp.pdf). Given that cod data were only used in IA analyses (see 'Tracing escapees' below), and the ability of markers to distinguish between populations provides a good indication of their power in IA analyses, the available SNPs for cod were ranked based on the pairwise F_{ST} values calculated among the wild cod populations using FSTAT v.2.9.3 (Goudet 1995). The Atlantic and Mediterranean sole data (wild and aquaculture) were used in both IA and PBT. To maximize the traceability power of selected sole SNPs for IA, markers were first ranked based on the pairwise F_{ST} values obtained from comparisons between the wild Atlantic and Mediterranean populations respectively. PBT analyses, on the other hand, require markers with a high genetic variability within a population to make robust assignments. Consequently, a second ranking of markers was based on their polymorphic information content (PIC) calculated with Cervus v.3.0 (Marshall et al. 1998) using the com-

Table 1. Available and newly generated single nucleotide polymorphism (SNP) genotypic datasets for wild and captive-bred populations of *Solea solea* and *Gadus morhua*. Atl-Aqua = aquaculture population sourced from the Atlantic Ocean, Med-Aqua = aquaculture population sourced from the Mediterranean Sea

Species	Origin	SNP genotypic data	
		Available	Source
Sole	Wild	427	Nielsen et al. (2012)
	Atl-Aqua	423	Diopere et al. (2014)
	Med-Aqua	181	Current study
Cod	Wild	1258	Nielsen et al. (2012)
	Atl-Aqua	427	Nielsen et al. (2012)

bined data from the respective broodstocks ($A_{\text{sole-BS}}$ and $M_{\text{sole-BS}}$) and their genetically similar wild populations (GER and ADR1 respectively). The top 96 ranking SNPs were used in all further analyses and further reduced genotypic datasets were used in the assignment analyses to determine the assignment power of the selected loci (Table 2).

This selection procedure for highly informative markers is unlikely to suffer from high-grading bias (Anderson 2010a, Waples 2010) for 3 reasons. First, assignment power was estimated from a different (holdout) set of samples to those used for SNP selection. Second, outlier SNPs were defined using initially high sampling sizes ($n \approx 40$) from various geographical locations, which reduces the effects of random sampling errors (Nielsen et al. 2012). Third, the use of 2 rigorous outlier detection methods and annotation information provides confidence that the high F_{ST} values of the selected markers are more likely to result from diversifying selection (i.e. real differentiation) rather than being at the extremes of a neutral marker F_{ST} distribution (Waples 2010, Nielsen et al. 2012).

Simulations of hatchery data

To formally evaluate the individual and combined impacts of F_n , N_e and the availability of reference data on the traceability efficiency of IA and PBT analyses, various breeding scenarios were simulated for both species. Data were simulated using the previously selected 96 SNPs of the $A_{\text{cod-BS}}$ and $M_{\text{sole-BS}}$ (including 7 individuals that died before reproduc-

tion) genotypes for cod and sole respectively (Fig. 1). An initial parental broodstock (P_{1-SIM}) and 4 offspring generations (F_{1-SIM} , F_{2-SIM} , F_{3-SIM} and F_{4-SIM}) were simulated under the assumption of perfect Hardy-Weinberg (HW) equilibrium. Different simulation series were performed using various N_e values ($N_e = 5, 10, 20$ or 50) to simulate drift due to reproductive variance. HYBRIDLAB v.1.0 (Nielsen et al. 2006) was used to simulate offspring genotypes used in the IA analyses. For the PBT analyses, Nookie v.1.0 (Anderson 2014) was used to simulate offspring genotypes because it generates individual genotypes 'bred' from specific parental pairs which are required for parentage assignment in simulated generations, rather than simply simulating individuals from a pool of population allele frequencies. Comparisons of genetic diversity showed that datasets generated through both programs were comparable (Supplement 3 at www.int-res.com/articles/suppl/q008p131_supp.pdf).

Comparative data analyses

A detailed comparison of the traceability results based on both the simulated and empirical datasets is important to determine the optimal traceability approach for a specific scenario. To be able to compare simulated and empirical results, population genetic parameters (F_{ST} values, observed heterozygosity H_{obs} and expected heterozygosity H_{exp}) associated with each dataset have to be understood as they will strongly influence the traceability power of the datasets. With the most comprehensive empirical data being available for the Mediterranean captive-

bred sole, population genetic parameters were calculated for the broodstock ($M_{\text{sole-BS}}$ and P_{1-SIM} [$N_e = 5, 10, 20, 50$]) and first-generation offspring ($M_{\text{sole-F1}}$ and F_{1-SIM} [$N_e = 5, 10, 20, 50$]). The Northern Adriatic population (ADR1), as the original source of $M_{\text{sole-BS}}$, was included in the analysis as a reference. Genetic diversity (H_{obs} and H_{exp}) was calculated for each independent dataset ($M_{\text{sole-BS}}$, $M_{\text{sole-F1-B1}}$, $M_{\text{sole-F1-B2}}$, $M_{\text{sole-F1-B3}}$, $M_{\text{sole-F1-B4}}$, P_{1-SIM} [$N_e = 5, 10, 20, 50$], F_{1-SIM} [$N_e = 5, 10, 20, 50$] and ADR1) using

Table 2. *Solea solea* and *Gadus morhua*. Datasets used in population (IA = individual assignment) and parentage (PBT = parentage-based tagging) analysis to test for effects of sampling regimes and traceability scenarios. Sampling Regime 1: reference data of the aquaculture population is available for the parental generation. Sampling Regime 2: reference data of the aquaculture population is limited to the founding broodstock. Scenario A: aquaculture broodstock originated from a genetically distinct population than the local wild populations. Scenario B: aquaculture broodstock originated from a local, genetically similar wild population. The number of single nucleotide polymorphisms (SNPs) used in each analysis is indicated between parentheses. SD = simulated data, EAD = empirical Atlantic data, EMD = empirical Mediterranean data, na = not applicable

Traceability method	Species	Sampling Regime		Scenario	
		1	2	A	B
IA	Sole	SD (96, 52)	SD (96)	na	EAD (96, 30, 1)
	Cod	SD (96)	SD (96)	na	na
PBT	Sole	SD (96, 48)	na	EAD (50, 30, 21) EMD (40, 35, 30)	EAD (50, 35, 30) EMD (40, 35, 30)
	Cod	na	na	na	na

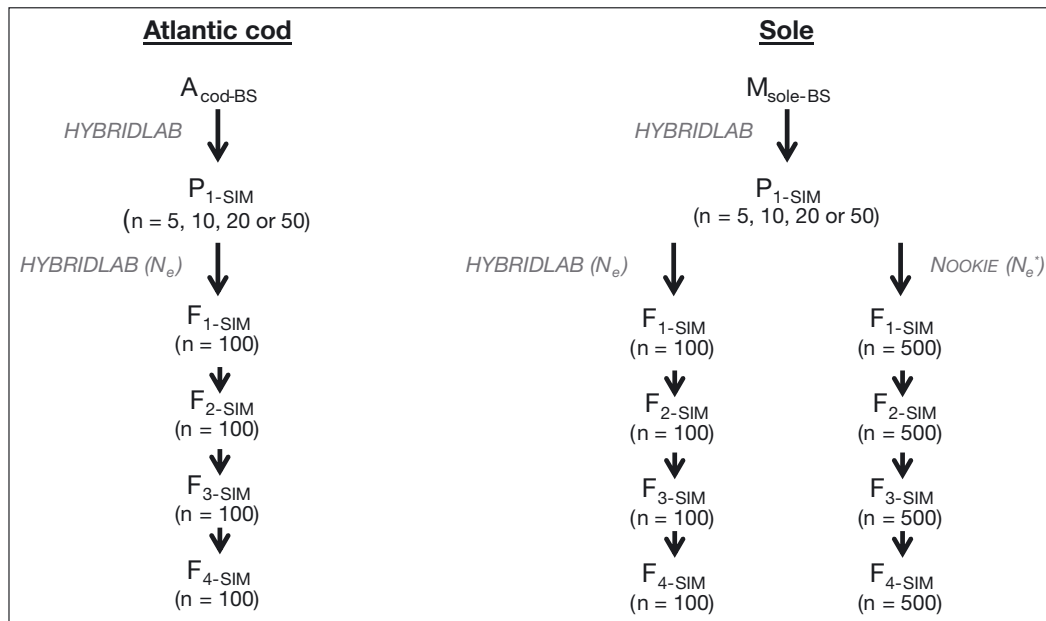


Fig. 1. *Gadus morhua* and *Solea solea*. Simulations used to generate captive-bred offspring genotypes. HYBRIDLAB v.1.0 (Nielsen et al. 2006) was used to generate the initial broodstock ($P_{1\text{-SIM}}$). Offspring genotypes used in individual assignment (IA) and parentage-based tagging (PBT) analyses were simulated with HYBRIDLAB and Nookie v.1.0 (Anderson 2014), respectively. $A_{\text{cod-BS}}$ = Atlantic cod broodstock, $M_{\text{sole-BS}}$ = Mediterranean sole broodstock, $F_{1\text{-SIM}}$, $F_{2\text{-SIM}}$, $F_{3\text{-SIM}}$, $F_{4\text{-SIM}}$ = 4 offspring generations, n = number of individuals, N_e (effective population size) = 5, 10, 20 or 50 and N_e^* = 4 or 50

Genetix v.4.05 (Belkhir et al. 2004). The realized levels of genetic differentiation between the simulated and empirical datasets was evaluated by calculating pairwise F_{ST} values and performing a discriminant analysis of principal components (DAPC) with the adegenet package in R v.3.0.2 (Jombart 2008, R Development Core Team 2010).

In addition to the comparative data analyses, the results also allow us to determine the N_e of the $M_{\text{sole-BS}}$. Using the H_{obs} and H_{exp} values obtained for $M_{\text{sole-F1-B1}}$, $M_{\text{sole-F1-B2}}$, $M_{\text{sole-F1-B3}}$ and $M_{\text{sole-F1-B4}}$, the N_e can be calculated for $M_{\text{sole-BS}}$ during each batch spawning event using the equation from Luikart & Cornuet (1999):

$$N_e = H_{\text{exp}} / [2(H_{\text{obs}} - H_{\text{exp}})] \quad (1)$$

Tracing escapees

Assignment efficiency is strongly influenced by the realized levels of genetic differentiation between the captive-bred and wild populations (IA) and the amount of genetic variability within the captive-bred population (PBT). By using the simulated datasets which are characterized by differences in the N_e and F_n , 2 parameters that significantly affect genetic differentiation and genetic variability, the effects of

these changes could be evaluated. In addition, the origin of the captive-bred population will also influence the traceability outcomes. To assess the effects of genetic dissimilarities between captive-bred and wild populations, 2 traceability scenarios were used: A, the broodstock originated from a genetically distinct population than the local wild populations; and B, the broodstock originated from a local, genetically similar wild population (Table 2).

From a forensic perspective, the ability to assign captive-bred fish back to their origin will be influenced by the nature and availability of reference samples, which may be challenging in well-established aquaculture species (Glover et al. 2009). For the purpose of our study, 2 simplified sampling regimes were used to evaluate the effect of missing data from previous captive-bred generations: Sampling Regime 1, in which data from the parental generation, which produced the escapees, is available, and escapees can thus be assigned to their parental generation or to the wild populations; Sampling Regime 2, in which data is restricted to the founding broodstock (often the case in operational hatcheries) and escapees can only be assigned to the founding broodstock or the wild populations. The lack of multiple captive-bred generations in the empirical data restricted the analyses of the empirical data to Sam-

pling Regime 1. Furthermore, PBT relies on the identification of parent–offspring relationships and will thus only be valuable under the assumptions of Sampling Regime 1.

IA and PBT analyses were performed using both simulated and empirical datasets (see Table 2). For the analyses of the simulated data, escapees were assumed to be flagged (i.e. genotypes of escapees are known) to obtain a baseline traceability efficiency, while for the analysis of the empirical data, escapees were mixed within a single wild population to create a more realistic scenario. IA analyses used the simulated datasets of both species and the empirical Atlantic sole data. PBT analyses were performed using sole data only, as Atlantic cod family data was unavailable.

IA analysis

IA analyses were performed with GeneClass2 v.2.0 using the ‘assign/exclude population as origin of individuals’ option (Piry et al. 2004). The threshold value was set to $p = 0.05$ and only individuals assigned to a population with rank 1 were considered. The probability of an individual being assigned to all possible reference populations was calculated using the Monte Carlo re-sampling method (Paetkau et al. 2004).

Using the simulated data of both species, assignment efficiency was evaluated under both sampling regimes. Input data for assignments consisted of 100 F_{n-SIM} genotypes (escapees) which could be assigned to either wild populations or their captive-bred population (i.e. their parental generation $F_{(n-1)-SIM}$ or their founding broodstock P_{1-SIM} for Sampling Regime 1 or 2 respectively).

For the analyses of the empirical data, 20 individuals from $A_{sole-F1}$ (10 from each full-sib family) representing the escapees were randomly selected and mixed with a genetically similar wild population (Scenario B) originating from the Belgian coast (BEL). Genotypes contained within this mixed population and the neighbouring wild populations (STO, GER, NOR, ENG, IS and GAS; see Supplement 1) could then be assigned to the remaining $A_{sole-F1}$ individuals.

PBT analysis

The parent–offspring relationships within the empirical aquaculture samples were obtained from

previous studies (Blonk et al. 2009) and additional parentage testing (Supplement 4 at www.int-res.com/articles/suppl/q008p131_supp.pdf). Only the SNP genotypes of individuals for which reliable parent–offspring relationships could be obtained were used in further analyses to ensure that the effectiveness of PBT could be formally evaluated. PBT analyses were performed with the software SNPPIT v.1.0 (Anderson 2010b), using only genotypic information (i.e. sex, age, year of sampling, etc. were considered unknown) and a genotyping error rate of 0.5 % per allele.

Using the simulated data and the wild populations of sole, the effect of N_e and F_n on the assignment success was evaluated under the assumption of Sampling Regime 1. Input files consisted of a list of putative parents (all wild populations and $F_{(n-1)-SIM}$) and offspring to be assigned (F_{n-SIM}) (i.e. the escapees).

Empirical analyses were performed with the Atlantic and Mediterranean sole data to determine the influence of the origin of the broodstock (Scenario A or B) on the traceability efficiency. Under Scenario A, the input file of putative parents contained the genotypes of $A_{sole-BE}$ or $M_{sole-BE}$ mixed with their respective source population (i.e. GER and ADR1 respectively). The offspring to be assigned contained a mixed population of 20 randomly selected $A_{sole-F1}$ or $M_{sole-F1}$ individuals added to genetically different wild populations (i.e. IS and THY respectively) and the remaining wild populations. In the case of Scenario B, assignment input was similar with the exception that the 20 randomly selected $A_{sole-F1}$ or $M_{sole-F1}$ individuals were mixed with a genetically similar wild population (i.e. BEL and ADR2 respectively).

RESULTS

Sampling and genotyping

Following complementary genotyping of the sole samples ($M_{sole-F1}$) with 192 SNPs, 181 SNPs passed the initial quality assessment. Of these, a panel of 96 highly informative SNP markers was selected and used in the analyses. An overview of all 96 selected SNPs used in the traceability analyses can be found in Supplement 2. Based on the re-genotyping of the 3 $A_{sole-BE}$ individuals at 181 loci, a genotyping discordance rate of 1.2 % was obtained. Hence, in all further analyses, a genotyping error of 1 % was used as an approximation.

Comparative data analyses

The comparative analyses of overall genetic diversity (H_{obs} and H_{exp}) showed no strong deviation between H_{obs} and H_{exp} in the $P_{1\text{-SIM}}$ ($N_e = 5, 10, 20, 50$) and $F_{1\text{-SIM}}$ ($N_e = 5, 10, 20, 50$) data (Fig. 2). However, in the $M_{\text{sole-F1}}$ data, a heterozygote excess was observed ($H_{\text{obs}} > H_{\text{exp}}$), suggesting that within $M_{\text{sole-F1}}$, a low number of individuals contributed to the next generation. Based on this heterozygote excess, the N_e is estimated to be 2.16, 2.10, 1.58 and 1.67 for $M_{\text{sole-F1-B1}}$, $M_{\text{sole-F1-B2}}$, $M_{\text{sole-F1-B3}}$ and $M_{\text{sole-F1-B4}}$ respectively.

Pairwise F_{ST} values and the DAPC show that both $M_{\text{sole-F1}}$ and ADR1 have a similar genetic composition (Fig. 3; Supplement 5 at www.int-res.com/articles/suppl/q008p131_supp.pdf). However, strong genetic differentiation is observed between $M_{\text{sole-F1}}$ and their population of origin ($M_{\text{sole-F1}}$ and ADR1). A comparison of the simulated data ($P_{1\text{-SIM}}$) and the wild populations ($M_{\text{sole-F1}}$ and ADR1) shows an increase in genetic differentiation when a strong bottleneck was applied (from $N_e = 50$ to $N_e = 5$), and the same pattern can be observed in the derived $F_{1\text{-SIM}}$ samples. Furthermore, the F_{ST} values are generally higher between the $M_{\text{sole-F1}}$ batches ($M_{\text{sole-F1-B1}}$, $M_{\text{sole-F1-B2}}$, $M_{\text{sole-F1-B3}}$, $M_{\text{sole-F1-B4}}$) than between the $F_{1\text{-SIM}}$ data (Fig. 3), and the same pattern can be observed with the DAPC (i.e. $F_{1\text{-SIM}}$ clusters are positioned closer together than $M_{\text{sole-F1}}$ clusters). One exception is the low genetic differentiation between $M_{\text{sole-F1-B3}}$ and $M_{\text{sole-F1-B4}}$, which is due to the same parents having produced these batches (Supplement 4). The results

suggest that the simulated data provides a good baseline (broodstock under HW equilibrium) for the validation of the traceability methods under real-life scenarios.

Tracing escapees

IA analysis

The success rate of correctly assigning escapees to the previous aquaculture generation (Sampling Regime 1) ranged from 73 to 100 % across all simulated datasets (Fig. 4). Results clearly indicate that the assignment success increased with increasing genetic drift (smaller N_e) and increasing generational distance from the original broodstock generation (higher F_n). Under the assumptions of Sampling Regime 2, the assignment success increased with increasing genetic drift, but no change in assignment success was observed with increasing generational distance from the broodstock (Fig. 4). However, in sole, an increasing F_n resulted in a decrease of assignment performance when a large effective population size ($N_e = 50$) was employed.

The population assignment analyses based on the empirical data of the Atlantic farmed sole and their neighbouring wild populations revealed that 81 % of escapees were correctly assigned using 1 SNP (average assignment score: 40), while a 100 % assignment was achieved with only 30 SNPs (average assignment score: 100).

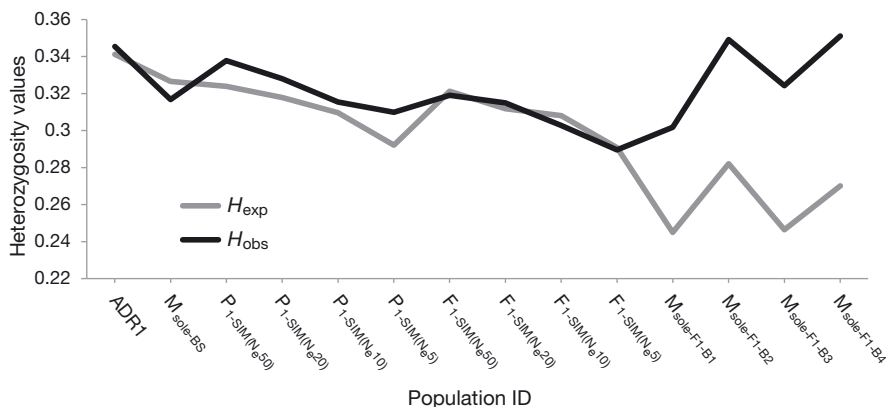


Fig. 2. Observed (H_{obs}) and expected (H_{exp}) heterozygosity values in population samples of Mediterranean sole *Solea solea*. ADR1 = natural population from Adriatic Sea; $M_{\text{sole-F1}}$ = broodstock composed of wild fish from Adriatic Sea; $P_{1\text{-SIM}}(N_e, X)$ ($X = 5-50$) = simulated parental populations at varying degrees of effective population size N_e ; $F_{1\text{-SIM}}(N_e, X)$ ($X = 5-50$) = simulated first offspring populations at varying degrees of effective population size N_e ; $M_{\text{sole-F1-BX}}$ ($X = 1-4$) = 4 actual offspring batches from aquaculture

PBT analysis

PBT analyses (SNPPIT) using the simulated data of sole showed that a panel of 48 SNP loci was sufficient to obtain an assignment success of $\geq 99\%$. Assignment success decreased (i.e. increasing number of non-excluded parent-offspring trios) with an increasing number of breeding generations (F_n), especially when N_e was small (Table 3).

The PBT results based on the empirical sole data show that under Scenario A, a dataset of 30 and 40 highly polymorphic SNPs was sufficient to trace back the Atlantic and Mediterranean aquaculture escapees, respectively (Table 4). Under the assumptions of Scenario B, a total of 35 highly poly-

M_{soleBS}	0.000																
M_{soleF1-B1}	0.078	0.000															
M_{soleF1-B2}	0.062	0.145	0.000														
M_{soleF1-B3}	0.057	0.161	0.080	0.000													
M_{soleF1-B4}	0.043	0.135	0.053	0.010	0.000												
P_{1-SIM}(N_e50)	0.002	0.068	0.061	0.050	0.040	0.000											
P_{1-SIM}(N_e20)	0.006	0.092	0.076	0.058	0.047	0.003	0.000										
P_{1-SIM}(N_e10)	0.009	0.085	0.076	0.061	0.048	0.005	0.007	0.000									
P_{1-SIM}(N_e5)	0.012	0.072	0.070	0.053	0.044	0.006	0.011	0.010	0.000								
F_{1-SIM}(N_e50)	0.002	0.048	0.046	0.035	0.029	0.002	0.003	0.003	0.003	0.000							
F_{1-SIM}(N_e20)	0.005	0.059	0.049	0.035	0.029	0.005	0.001	0.003	0.003	0.007	0.000						
F_{1-SIM}(N_e10)	0.007	0.058	0.056	0.043	0.036	0.008	0.004	0.000	0.002	0.008	0.008	0.000					
F_{1-SIM}(N_e5)	0.017	0.070	0.074	0.053	0.047	0.020	0.012	0.005	0.000	0.022	0.022	0.016	0.000				
ADR1	0.010	0.070	0.058	0.062	0.050	0.011	0.015	0.015	0.013	0.011	0.016	0.021	0.030	0.000			
	M_{soleBS}	M_{soleF1-B1}	M_{soleF1-B2}	M_{soleF1-B3}	M_{soleF1-B4}	P_{1-SIM}(N_e50)	P_{1-SIM}(N_e20)	P_{1-SIM}(N_e10)	P_{1-SIM}(N_e5)	F_{1-SIM}(N_e50)	F_{1-SIM}(N_e20)	F_{1-SIM}(N_e10)	F_{1-SIM}(N_e5)	ADR1			

Fig. 3. Pairwise F_{ST} values matrix among the simulated and empirical datasets of Mediterranean aquaculture sole *Solea solea*. □: $F_{ST} = 0$; □: $0 < F_{ST} < 0.01$; □: $0.01 \leq F_{ST} < 0.05$; □: $0.05 \leq F_{ST} < 0.1$; □: $0.1 \leq F_{ST}$. See Fig. 2 for abbreviations

morphic SNPs were sufficient for the identification of all aquaculture escapees for both broodstocks (Table 4).

DISCUSSION

Our study shows that a panel of highly informative, gene-associated SNP markers can discriminate between wild and captive-bred marine fish, even without extensive domestication of the species of interest. Furthermore, the results show that IA and

PBT analyses can both be valuable tools for wildlife forensics and fisheries management, depending on the genetic history of the relevant captive populations.

Potential of SNP markers for traceability

Biallelic SNP markers are generally considered less informative than microsatellite markers. However, SNPs are highly abundant and evenly distributed throughout the genome (Morin et al. 2004).

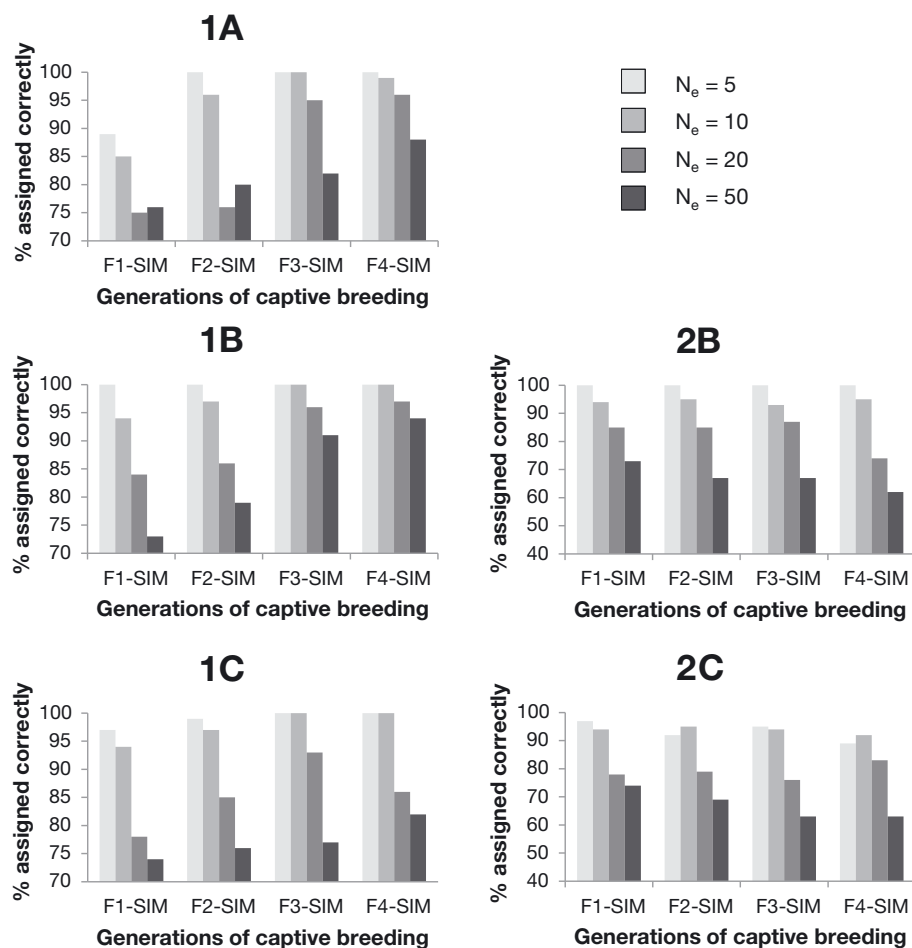


Fig. 4. *Gadus morhua* and *Solea solea*. Percentage of correctly assigned individuals in individual assignment (IA) analysis using simulated datasets. 1A, 1B, 1C: Sampling Regime 1 (reference data of aquaculture population is available for parental generation); 2B, 2C: Sampling Regime 2 (reference data of aquaculture population is limited to the founding broodstock); with results based on (A) sole data using 52 single nucleotide polymorphisms (SNPs), (B) sole data using 96 SNPs, (C) cod data using 96 SNPs. F_{1-SIM}, F_{2-SIM}, F_{3-SIM}, F_{4-SIM} = 4 offspring generations, N_e = effective population size

Hence, low polymorphism levels can be compensated through the development of a large number of gene-associated SNPs which can detect even small population genetic differences (Nielsen et al. 2012). Additionally, SNP genotyping can be highly automated and does not require extensive calibrations for marker exchange (Hauser & Seeb 2008). These characteristics make SNP markers ideal for the development of universally applicable genetic traceability tools, which inherently rely on the availability of robust reference data (Helyar et al. 2011, Nielsen et al. 2012). In the case of tracing captive-bred marine fish, SNPs can be used to detect subtle genetic differences between wild and captive-bred populations, even after just a few generations of captive breeding. Consequently, there is ample opportunity to use SNP-based tracing in fisheries management and wildlife forensics. From a management perspective, SNPs can be employed to monitor the effects of accidental/deliberate releases of captive-bred fish on wild populations. SNP-based tracing will also have forensic applications, as it will be a useful tool in the

fight against mismanagement practices in aquaculture and the mislabelling of seafood products, since universal markers for the identification of captive-bred individuals can be developed (Karlsson et al. 2011).

Applications of IA and PBT analyses

Our results demonstrate that IA and PBT perform optimally under different scenarios. The performance of IA analyses improves with increased genetic differentiation between the aquaculture and wild populations as a result of increased generational breeding (high F_n) and/or a low N_e in the broodstock. PBT analyses, on the other hand, perform better when a high N_e characterizes the broodstock and/or generational breeding is low. This is as expected, given that candidate parents are less likely to be excluded from being the real parents due to loss of genetic diversity (low N_e and/or high F_n). As a result of the performance differences, the suitability of IA

Table 3. *Solea solea*. Parentage-based tagging (PBT) analysis using software package SNPPIT to identify escapees based on simulated sole data. F_{n-SIM} = number of captive-bred generations that were simulated; F_{1-SIM} , F_{2-SIM} , F_{3-SIM} , F_{4-SIM} = 4 offspring generations that were simulated; N_e = effective population size; na = not applicable

Loci	N_e	F_{n-SIM}	% assigned to correct population	Proportion assignments with $p > 0.05$	Number of non-excluded parentage trios ($\times 10^3$)	Number of non-excluded trios from the wrong population ($\times 10^3$)
96	50	F_{1-SIM}	100	0.00	0.69	0.05
		F_{2-SIM}	100	0.04	4.98	0.07
		F_{3-SIM}	100	0.04	5.95	0.09
		F_{4-SIM}	100	0.04	7.70	0.13
	4	F_{1-SIM}	100	0.00	0.54	0.04
		F_{2-SIM}	100	0.34	471.06	0.18
		F_{3-SIM}	100	0.28	400.90	1.84
		F_{4-SIM}	na	na	na	na
	50	F_{1-SIM}	99	0.01	6.65	1.94
		F_{2-SIM}	100	0.11	81.82	5.18
		F_{3-SIM}	99	0.10	122.12	8.41
		F_{4-SIM}	100	0.25	240.94	10.69
48	4	F_{1-SIM}	100	0.00	1.72	1.14
		F_{2-SIM}	99	0.11	81.83	5.18
		F_{3-SIM}	na	na	na	na
		F_{4-SIM}	na	na	na	na

and PBT analyses is strongly dependent on the ultimate goal of genetic tracing studies. Hence, our results are important for wildlife forensics and fisheries management to determine the optimal assignment strategy.

A common goal of fisheries management is the preservation or restoration of commercially important fish populations to levels which will produce a long-term maximum sustainable yield (MSY) (FAO Fisheries and Aquaculture Department 2008). Since the number of overexploited marine fish populations con-

tinues to increase, stock enhancement and sea ranching programmes have become popular management actions (Bell et al. 2008, FAO Fisheries and Aquaculture Department 2012). Consequently, the release of first-generation captive-bred juvenile fish which are genetically similar to the local wild populations has increased (Bell et al. 2008). Given that PBT analyses have a high identification efficiency for first-generation escapees and can detect hybridization between wild and captive-reared conspecifics, they can be used to jointly evaluate the levels of introgression (enforcement action) and the efficiency of restocking, stock enhancement and sea ranching programmes (management action).

Robust, forensically validated and universally applicable traceability tools can also be used in wildlife forensics to support legal actions against mismanagement of aquaculture facilities, which increases the chance of escapees, or the mislabelling of seafood products for financial profits (Ogden 2008, Glover 2010, Hanner et al. 2011). Our results indicate that both IA and PBT are potentially valuable provided that the aquaculture history of the species of interest is taken into account. IA analyses are a powerful tool for species with a long aquaculture history since cap-

Table 4. *Solea solea*. Parentage-based tagging (PBT) approach using software package SNPPIT for identifying escapees based on the empirical sole aquaculture data. Scenario A: broodstock originated from a genetically different population than the local wild populations, Scenario B: broodstock originated from a local wild population. SNP = single nucleotide polymorphism

Broodstock origin	Scenario	Number of SNPs	Escapees		Natural individuals	
			% assigned to both parents	% significantly assigned	% assigned to at least 1 parent	% significantly assigned
Atlantic Ocean	A	50	100	100	6	0
	A	30	100	100	21	0
	A	21	85	35	26	2
	B	50	100	100	6	0
	B	35	100	100	15	0
	B	30	100	95	19	0
Mediterranean Sea	A	40	100	100	52	0
	A	35	100	95	51	0
	A	30	95	80	74	0.7
	B	40	100	100	84	0
	B	35	100	100	81	0
	B	30	95	80	90	0

tive breeding has resulted in a strong genetic differentiation between captive-bred and wild fish populations (Bekkevold et al. 2006, Karlsson et al. 2011). However, most marine fish species have only recently been bred in captivity and thus forensic tools need to be able to differentiate between genetically similar captive-bred species and wild conspecifics. Our findings suggest that PBT can be used for these recently domesticated fish species, since assignment success was high after only a single generation of captive breeding. This is in line with expectations, since PBT was originally developed to identify the source of salmon released into rivers and is thus capable of differentiating between genetically similar hatchery populations (Anderson & Garza 2006). Genetic assignment methods have already been successfully applied in a forensic context (Wong & Hanner 2008, Glover 2010). However, real-life situations often complicate genetic tracing studies (Glover et al. 2009). As such, the presence of multiple (genetically similar) putative source farms and the lack of extensive genetic reference data will reduce the assignment efficiency of both IA and PBT. Although the latter is less problematic for IA analysis, PBT unequivocally requires genotypic information from all parental individuals that have contributed to the subsequent generation. The increased use of genetic broodstock management and selective breeding programmes might partially resolve this but the feasibility of using PBT in a forensic context remains controversial (Blonk et al. 2010, Vandeputte et al. 2011).

Validation of traceability approaches

Validating traceability methods requires a detailed comparison between expected (simulations) and observed (empirical) results. The assignment success rates of the analyses based on the F_{1-SIM} and the empirical data reveal that overall, a higher success rate is obtained in the empirical analyses. The fact that relatively more SNP makers are needed for unambiguous assignments in the simulated data can be explained by a high reproductive skew in real aquaculture production ($A_{sole-BS}$ and $M_{sole-BS}$), which is difficult to simulate with currently available software packages. From the N_e values estimated based on the observed heterozygote excess in the $M_{sole-F1}$, we conclude that on average, 2 parental individuals contributed to each offspring batch, and these findings are supported by the results from the additional parentage testing (Supplement 4). Furthermore, comparing the genetic differentiation between the em-

pirical and simulated data (DAPC) suggests that within the $M_{sole-BS}$, an N_e of between 5 and 10 is the most likely, which is supported by the N_e estimates found in the $A_{sole-BS}$ by Blonk et al. (2009).

Other evidence supporting the methodology employed here arises from the comparison of current results with earlier studies. Vandeputte et al. (2011) recorded a decrease in the assignment power when comparing theoretical, simulated and empirical parentage assignments using microsatellite data. However, our study has clearly indicated that large-scale SNP genotyping (i.e. genome scan) combined with a selection procedure for highly informative gene-associated markers (high F_{ST} values and PIC) can increase the assignment power in empirical studies. This is consistent with the findings of previous studies which recorded similarly high assignment efficiencies with only a small number of markers (Nielsen et al. 2012). Hence, the methodology presented here will be valuable for future traceability studies where sufficient genetic background information is available for the species of interest. With low-cost high-throughput genotyping-by-sequencing methods now available to be implemented in breeding programmes (Davey et al. 2011), the cost of developing a large battery of markers should not impede applications to fisheries management and wildlife forensics.

CONCLUSIONS

This study has evaluated the relative power of parentage-based tagging (PBT) and individual assignment (IA) for identifying the population of origin of marine aquaculture fish under a range of scenarios, highlighting the benefits and disadvantages of each. PBT potentially offers the strongest line of traceability evidence, as the identification of a specific parental pair with high confidence is likely to be more powerful than a combined population assignment and exclusion approach under IA, particularly where aquaculture and wild populations have not diverged significantly. The results presented here have shown that PBT analyses will be particularly valuable in fisheries management to evaluate the genetic effects and the impact of accidental and/or deliberately released captive-bred fish. However, current aquaculture practices restrict the practical application of PBT due to the requirement for complete broodstock sampling; consequently, in most marine fish aquaculture scenarios, IA analyses are considered to be of more practical use for future

traceability applications. Ultimately, the availability of genetic background information and the aim of the study will determine whether IA or PBT will be the method of choice.

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